

Variational Monte Carlo:

$$H\psi = E\psi \quad \psi(\vec{R}) \quad \vec{R} = (\vec{r}_1, \vec{r}_2, \dots, \vec{r}_N)$$

only interested in  $E_0 + \psi_0$  ground state

$$E_V = \frac{\int \psi^* H \psi d\vec{R}}{\int |\psi|^2 d\vec{R}} > E_0$$

$$= \frac{\int (H\psi)/\psi (\psi\psi^*) d\vec{R}}{\int |\psi|^2 d\vec{R}}$$

$$E_L(\psi) = \frac{H\psi}{\psi}$$

local energy

if  $\psi \rightarrow \psi_0$

$E_L(\psi) \rightarrow E_0$   
const.

$$= \int E_L(\vec{R}) P(\vec{R}) d\vec{R} \quad P(\vec{R})$$

$$P(\vec{R}) = \frac{|\psi(\vec{R})|^2}{\int |\psi|^2 d\vec{R}}$$

$$E_V = \langle E_L \rangle = \frac{1}{n} \sum_{i=1}^n E_L(\vec{R}_i)$$

with  $\vec{R}_i \in P(\vec{R})$



By the Metropolis  
algorithm

How to choose  $\psi$ ? the trial ground state

$$V(r) = -\frac{Z}{r} \quad V_{IJ}(r) = \epsilon_0 \left( \frac{1}{r_{I2}} - \frac{1}{r_{I1}} \right) \text{ etc.}$$

↳ short-range, hard-core

$$\rightarrow -\frac{\hbar^2}{2m} \nabla^2 \psi + V(r) \psi = E \psi \quad \text{potential}$$

$m = \text{reduced mass}$

$$\text{Schrödinger form: } -\frac{\hbar^2}{2m} \nabla^2 \psi + V(r) \psi = E \psi$$

choose  $r_0$  basic unit of length.

$$r = r^* r_0$$

$$-\frac{\hbar^2}{2m} \left( \frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2} + \frac{\partial^2}{\partial z^2} \right) \psi \dots$$

$$-\frac{\hbar^2}{2m} \frac{1}{r_0^2} \left( \frac{\partial^2}{\partial x^{*2}} + \dots \right) \psi \dots = E \psi$$

$$E_0 = -\frac{\hbar^2}{2m} \frac{1}{r_0^2} \quad + V \psi = E \psi$$

$$V = V^* E_0$$

$$-\frac{1}{2} \cancel{E_0} \nabla^2 \psi + \frac{V}{\cancel{E_0}} \psi = \cancel{E^*} \cancel{E_0} \psi$$

$$\boxed{-\frac{1}{2} \nabla^2 \psi + V^* \psi = E^* \psi}$$

choose  $r_0$  such that  $V^*$  is the simplest!

$$\text{if } V^* = -\frac{Z}{r} \quad \text{if } r_0, E_0$$

are the atomic

$$E_0 = -\frac{1}{2} \frac{1}{n^2} \text{ Hartree units.}$$

$$= -\frac{1}{n^2} R_b \quad \text{Hartree} = 2 R_b,$$

$$-\frac{1}{2\mu} \nabla^2 \psi + V(r) \psi = E \psi \quad \text{for the}$$

$$g.s. \psi > 0$$

$$\psi = e^{-S(r)}$$

$$\bar{\psi} = -\bar{\nabla} S e^{-S(r)}$$

$$-\frac{1}{2\mu} [-S''(r) - \frac{2}{r} S'(r) + (S')^2] \psi \quad \bar{\nabla}^2 \psi = [-\nabla^2 S + (\nabla S)^2] \psi$$

$$+ V(r) \psi = E \psi \quad = \left[ -\frac{\hbar^2}{2\mu} S'' + \left( \frac{\hbar^2}{2\mu} \right)^2 \bar{\nabla} S = S' \frac{\hbar^2}{2\mu} \right]$$

$$= \frac{1}{2\mu} S'' + \frac{1}{\mu} \frac{S'}{r} - \frac{1}{2\mu} (S')^2 + V(r) = E \quad \text{Evid. } E_f$$

$E_L$

must  
cancel the

singularity in  $V(r)$

$\downarrow$  const

$$-\frac{\hbar^2}{2\mu} \text{ or } E \left[ \frac{1}{r_2} - \frac{1}{r_1} \right]$$

$$\text{take } S = ar^\eta \quad S' = a\eta r^{\eta-1}$$

$$S'' = a\eta(\eta-1)r^{\eta-2}$$

$$\frac{1}{2\mu} a\eta(\eta-1)r^{\eta-2} + \frac{1}{\mu} a\eta r^{\eta-2} + \frac{1}{2} \frac{\hbar^2(\eta-1)}{r} + \eta = \frac{1}{2} \frac{\hbar^2(\eta+1)}{r}$$

$$-\frac{1}{2\mu} a^2 \eta^2 r^{2\eta-2} + V(r)$$

for LJ

$\downarrow$  = const

$$r^{-12}$$

$$2\eta-2 = -12$$

$$2\eta = -10$$

$$\eta = -5$$

$$S = ar^{-5} !$$

$$\text{take } \mu = \frac{1}{2} \quad \frac{1}{2\mu} a^2 25 = E$$

$$a = \frac{\sqrt{E}}{5} !$$

$$\eta = 1 \quad \eta-2 = -1$$

$$\frac{1}{2} a(\eta)(\eta+1) \quad \eta = 1 !$$

$$a = 2$$

$$S = 2r ! \quad \text{cusp condition!}$$

$$E_L = \frac{H\psi}{\psi} = \frac{H e^{-S}}{e^{-S}}$$

=

$$\Psi(r) = e^{-\sum_i s_1(r_i) - \underbrace{\sum_{i,j} s_2(r_{ij})}_{\text{two-particle}} - \sum_{i,j,k} s_3(r_{ijk})}$$

Det

↑  
functions

$s_1(r_i) = z r_i$   
 Correlated basis function expansion.  
 → MC calculations.

only for  
Bosons

## More probability theory: stochastic process

1)  $P(x)$  pdd for  $X$

2)  $P(x, y)$  joint pdd. for  $X$  &  $Y$   $P(y) = \int dx P_{XY}(x, y)$   
 $P(x) = \int dy P_{XY}(x, y)$

3)  $P_{XY}(x, y) = P_{X|Y}(x, y) P(y)$   
 $\uparrow$  conditional prob.  
 conditional on  $Y$ .  
 prob. on  $x$ , depending on  $y$ .

4)  $P_N(x_1, x_2, \dots, x_N) = P(x_N | x_{N-1}, \dots, x_1; x_{N-1}, \dots, x_1)$

joint prob.  
for  $N$  particles  
let  $x_1, x_2, \dots, x_N$   
be  $X(dt), X(2dt), \dots, X(Ndt)$ .

$P(x_{N-1} | x_{N-2}, \dots, x_1)$   
 $P(x_{N-2} | \dots)$   
 $P(x_2 | x_1) P(x_1)$

time-interpretation  $\Rightarrow$  stochastic process

$X(t)$

||

$P(x, t)$

$\uparrow$   
different types of

$P_N(x_1, \dots, x_N) = P_N(x_N) P_{N-1}(x_{N-1}) \dots P_1(x_1)$   
 $P_1(x_1)$

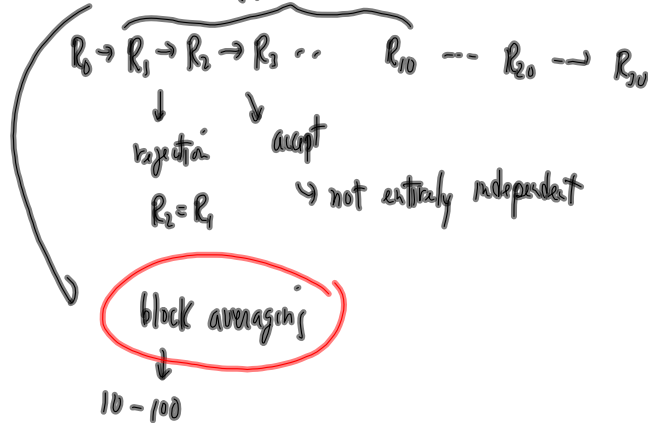
total random process.

$P_N(x_1, \dots, x_N) = P(x_N | x_{N-1}) P(x_{N-1} | x_{N-2})$   
 $\dots P(x_2 | x_1) P(x_1 | x_1) P(x_1)$

Markov process.

$$\langle f(\vec{R}) \rangle = \frac{1}{N} \pm \frac{\sigma_f}{\sqrt{N}} \quad \sigma_f^2 = \langle f^2 \rangle - \langle f \rangle^2$$

$n = n$ -independent configuration  
to  $\sum_{i=1}^n f(\vec{r}_i)$



Only Cartesian coordinates for many-particles.

$$|\psi(\vec{R})|^2 dx_1 dy_1 dz_1 dx_2 dy_2 dz_2 \dots$$

~~$$|\psi(r)|^2 4\pi r^2 dr$$~~

$$= |\psi(r)|^2 dx dy dz$$

$$e^{-\frac{1}{2\alpha t} x^2} |\psi|^2 \quad r = \sqrt{x^2 + y^2 + z^2}$$

$$= e^{-2\alpha r^2} = e^{-2\alpha(x^2 + y^2 + z^2)}$$

$$2\alpha = \frac{1}{2\alpha t} \quad \alpha =$$

$$\alpha t = \frac{1}{4} \frac{1}{\alpha} \quad \checkmark$$

$$g \propto \sqrt{\alpha t} \quad g \perp$$

Markov process :

$$P(x_N, x_{N-1}, \dots, x_2, x_1) = P(x_N, x_{N-1}) P(x_{N-1}, x_{N-2}) \dots P(x_2, x_1) P(x_1)$$

$$P(x_N) = \int dx_{N-1} \dots dx_1 \underbrace{P(x_N, x_{N-1}) \dots P(x_1)}_{\text{Path-Integral}} P(x_1)$$

final prob. dist

$$= \int dx_{N-1} P(x_N, x_{N-1}) P(x_{N-1}) \quad \hookrightarrow \text{in imaginary time.}$$

$$P(x') = \int dx \underbrace{P(x', x)}_{\text{conditional prob.}} P(x) \quad \vec{P}' = M \vec{P}$$

$M$   $\nearrow$

$= \text{prob. dist. for } x', \text{ given } x$

$$\vec{P}_{n+1} = (M)^N \vec{P}_1$$

Choose  $P(x', x)$  such that

$$P_N(x) \xrightarrow{N \rightarrow \infty} P_e(x) \quad \nwarrow \text{equilibrium}$$

$$\sim |\psi|^2$$

How to choose  $P_2(x', x)$  such that

$$P_N(x) \xrightarrow{N \rightarrow \infty} P_e(x)$$

Requirement on  $P(x', x)$ :

i)  $P(x', x) \geq 0$

ii)  $\int dx' P(x', x) = 1$

$$P_{n+1}(x') = \int dx P(x', x) P_n(x)$$

$$1 \cdot 1 = \int dx' 1 P(x', x)$$

1 is the left eigenvector of  $M$  with eigenvalue = 1

at equilibrium

$$P_e(x') = \int dx P(x', x) P_e(x)$$

$P(x', x)$   $\neq$  symmetric

Stationary condition  $\rightarrow$

$$\vec{P}_e = M \vec{P}_e$$

$P_e(x)$  is the right eigenvector of  $M$  with eigenvalue = 1

In general

$$\int dx P(x', x) R_n(x) = \lambda_n R_n(x)$$

$$\int dx' L(x') P(x', x) = \lambda_n L_n(x)$$

$$P(x', x) = \sum_n \lambda_n L_n(x') R_n(x)$$



All  $\lambda_n$  are  $< 1$

"Proof" Let  $C$  be the max  $|L_n(x)|$

$$\left| \int dx' L_n(x') P(x', x) \right| < C \left| \int dx' P(x', x) \right|$$

"1"

$$|\lambda_n L_n(x)| < C$$

$$|\lambda_n| < \frac{C}{|L_n(x)|} < 1$$

↑  
choose to be  $C$

How to choose  $p(x', x)$  ?

$$p(x', x) = \underbrace{w(x', x)}_{\substack{\uparrow \\ \text{conditional} \\ \text{prob.}}} + \underbrace{\left[ 1 - \int w(x', x) dx' \right]}_{\substack{\text{moving} \\ \text{prob.}}} \delta(x', x)$$

↑ satisfy (i)

$\int w(x', x) dx' < 1$

Prob. of NOT moving.

$$p_e(x') = \int p(x', x) p_e(x) dx$$

$$= p_e(x') + \int w(x', x) p_e(x) dx - \int dx'' w(x'', x') p_e(x')$$

$\parallel$   $\uparrow$   
 $dx$   $x$

$$p_e(x') = p_e(x') + \int dx \left[ w(x', x) p_e(x) - w(x, x') p_e(x') \right]$$

choose  $w(x', x)$  such that

$$w(x', x) p_e(x) = w(x, x') p_e(x')$$

detailed-balance,  
condition

Given  $\underline{X}, \underline{Y}$  described by  $P_{\underline{XY}}(x, y)$

What is  $P_z(z)$   $z = \underline{X} + \underline{Y}$ ?  $\hookrightarrow P_{\underline{X}}(x) P_{\underline{Y}}(y)$

$$P_z(z) = \int dx dy \delta(z - x - y) P_{\underline{XY}}(x, y)$$

$$= \int dx P_{\underline{XY}}(x, z - x)$$

$$P_z(z) = \int dx P_{\underline{X}}(x) P_{\underline{Y}}(z - x)$$

$$\text{Let } \underline{Y} = \Delta \underline{X}$$

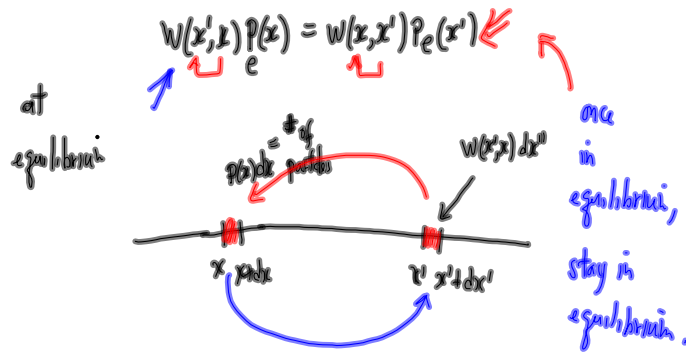
$\underbrace{\hspace{1cm}}$   
convolution

$$z = \underline{X} + \Delta \underline{X}$$

$$P_1(x') = \int dx P_1(x) P_2(x', x) \stackrel{= \Delta x}{\hookrightarrow} P(x' - x)$$

$$\underline{X}' = \underline{X} + \Delta \underline{X} \hookrightarrow \text{given by } P_2(\Delta x)$$

Detail-balance condition:



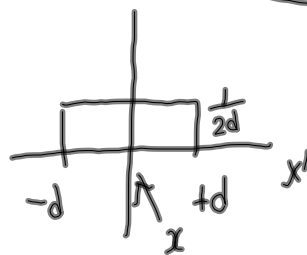
to devise a  $w(x, x')$

$$w(x', x) = T(x', x) A(x', x)$$

$T$  trial move prob.  
 $A$  acceptance prob.

take  $T(x', x)$  to be symmetric

$$T(x', x) = \frac{1}{2d}$$

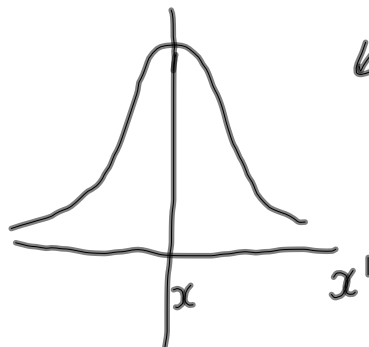


$$\text{if } |x' - x| < d$$

$$= 0 \quad |x' - x| > d$$

$$x' = x + d * (2 * \text{ran} - 1)$$

$$\text{or } T(x', x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2} (x' - x)^2}$$



$$x' = x + \sqrt{\sigma^2} g$$

$g$  Gaussian RV.

$$W(x', x) P_e(x) = W(x, x') P_e(x')$$

$$\text{||} \\ \cancel{T(x', x)} A(x', x) P_e^{(x)} = \cancel{T(x, x')} A(x, x') P_e(x')$$

$$\text{let } P_e(x) = e^{-S(x)}$$

$$A(x', x) e^{-S(x)} = A(x, x') e^{-S(x')}$$

$$A(x', x) = A(x, x') e^{-\Delta S} \quad \Delta S = S(x') - S(x)$$

choose  $A(x', x)$  to be a function just of  $\Delta S(x', x)$

$$A(\Delta S) = A(-\Delta S) e^{-\Delta S}$$

$$A(\Delta S) e^{\frac{1}{2}\Delta S} = A(-\Delta S) e^{-\frac{1}{2}\Delta S}$$

||  
 $F(\Delta S)$  is an Even function of  $\Delta S$ !

$$\begin{aligned} A(\Delta S) &= F(\Delta S) e^{-\frac{1}{2}\Delta S} \quad \text{if} \\ &= e^{-\frac{1}{2}|\Delta S|} e^{-\frac{1}{2}\Delta S} \quad \Delta S < 0 \\ &\quad \text{max} \end{aligned}$$

then if  $\Delta S > 0$   $A = 1$

$$= e^{-\Delta S} = \frac{P_e(x')}{P_e(x)}$$

$$A(\Delta S) = \begin{cases} 1 & \text{if } \Delta S < 0 \\ e^{-\Delta S} & \text{if } \Delta S > 0 \end{cases}$$

$$= \min(1, e^{-\Delta S})$$

## Generalization of Metropolis

Let now take  $T(x', x)$  **NOT** symmetric

$$T(x', x) = \frac{1}{\sqrt{2\pi\Delta t}} e^{-\frac{1}{2\Delta t} (x' - x - v(x)\Delta t)^2}$$

$x' = x - v(x)\Delta t + \sqrt{\Delta t} \, g$

↑ with a drift velocity field.

$$T(x', x) A(x', x) = T(x, x') A(x, x') e^{-\Delta S} \quad \text{Langevin algorithm}$$

$$A(x', x) = A(x, x') \frac{T(x, x')}{T(x', x)} e^{-\Delta S}$$

$$= A(x, x') e^{-\tilde{\Delta S}} \quad e^{-\tilde{\Delta S}} = \frac{T(x, x')}{T(x', x)} e^{-\Delta S}$$

$$A(\tilde{\Delta S}) = \min \left( 1, e^{-\tilde{\Delta S}} \right)$$

$$= \min \left( 1, \frac{T(x, x') p_q(x')}{T(x', x) p_q(x)} \right) \quad \mathcal{R}$$

Monte Carlo  $\rightarrow$  sampling  $P_e(x) = P_e(\vec{R})$

- 1) Inversion in 1D
- 2) Metropolis in the general case:  $\vec{R}$   $\rightarrow$
- 3) Langevin algorithm:

Consider the diffusion equation (imaginary time Schrödinger Eq.)

$$-\frac{\partial \psi}{\partial t} = -\frac{1}{2} \nabla^2 \psi$$

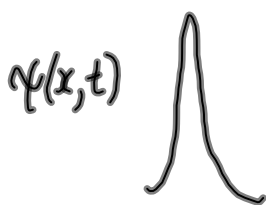
$$\psi(x', t+\delta t) = \int G(x', x, \delta t) \psi(x, t) dx$$

$$G(x', x; \delta t) = \frac{1}{\sqrt{2\pi\delta t}} e^{-\frac{1}{2\delta t}(x'-x)^2}$$

$$\vec{x}' = \vec{x} + \sqrt{\delta t} \vec{g}$$

if each particle  
undergoes Gaussian  
random walk  
their "density"

as  $t \rightarrow \infty$



$t=0$



uniform equilibrium  
distribution

To get at a non-trivial  $\rho_c(x)$ ,

$$\rho(x,t) = \phi(x)\psi(x,t) \quad -\frac{\partial \psi}{\partial t} = -\frac{1}{2}\nabla^2 \psi$$

$\uparrow$   $\searrow$  const  
 distribution is constant  
 $\rho(x,t) \rightarrow \text{const } \phi(x)$   
 $t \rightarrow \infty$

$$-\frac{\partial \rho}{\partial t} = \phi \left( -\frac{\partial \psi}{\partial t} \right) = -\frac{1}{2} \nabla^2 \psi \phi \quad \checkmark \text{ express in terms of } \rho$$

$$\begin{aligned} \phi \nabla^2 \psi &= \vec{\nabla}(\phi \vec{\nabla} \psi) - \underbrace{\vec{\nabla} \phi \cdot \vec{\nabla} \psi}_{-\nabla(\psi \nabla \phi) + \psi \nabla^2 \phi} \\ &= \vec{\nabla}(\phi \vec{\nabla} \psi - \psi \nabla \phi) + \phi \psi \left( \frac{\nabla^2 \phi}{\phi} \right) \\ &= \vec{\nabla}(\phi \vec{\nabla} \psi - \psi \nabla \phi) + \underbrace{\phi \psi \left( \frac{\nabla^2 \phi}{\phi} \right)}_{\vec{\nabla} \cdot (\rho \frac{\vec{\nabla} \phi}{\phi})} \\ &= \nabla^2 \rho - 2 \vec{\nabla} \cdot \left( \rho \frac{\vec{\nabla} \phi}{\phi} \right) + \rho \left( \frac{\nabla^2 \phi}{\phi} \right) \end{aligned}$$

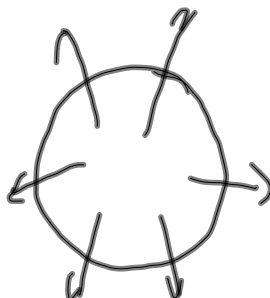
$$-\frac{\partial \rho}{\partial t} = -\frac{1}{2} \nabla^2 \rho + \vec{\nabla} \cdot \left( \rho \frac{\vec{\nabla} \phi}{\phi} \right) + \left( -\frac{1}{2} \frac{\nabla^2 \phi}{\phi} \right) \rho$$

We  $-\frac{\partial \rho}{\partial t} = -\frac{1}{2} \nabla^2 \rho$  diffusion

$$-\frac{\partial \rho}{\partial t} = \vec{\nabla} \cdot (\rho \vec{v}) \quad \vec{v} = \frac{\vec{\nabla} \phi}{\phi}$$

$\uparrow$   
equation of continuity

$J = \rho \vec{v}$   
current density





$$-\frac{\partial}{\partial t} \rho = -\frac{1}{2} \nabla^2 \rho + \nabla \cdot (\rho \vec{v})$$

Fokker-Planck equation.

$$0 = \nabla \cdot (-\frac{1}{2} \nabla \rho + \rho \vec{v})$$

$$\frac{\nabla \rho}{\rho} = 2 \vec{v} = 2 \frac{\nabla \phi}{\phi}$$

$$\rho(x,t) \xrightarrow{t \rightarrow \infty} \rho_e = \phi^2$$

Can simulate  $\Delta t$

$$x' = x + \sqrt{\Delta t} g$$

$$\frac{dx}{dt} = v(x)$$

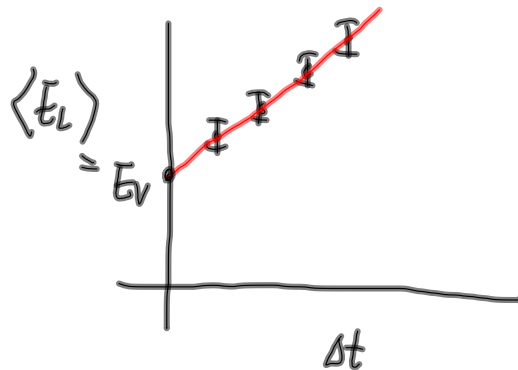
$$x' = x + v(x) \Delta t$$

$$x' = x + v(x) \Delta t + \sqrt{\Delta t} g$$

$$-\frac{\partial \rho}{\partial t} = \nabla \cdot (\rho \vec{v})$$

first order Langevin algorithm.

$$\vec{v} = \frac{\nabla \phi}{\phi}$$



$$\text{if } \phi = e^{-\alpha r}$$

$$\vec{v} = \frac{\nabla \phi}{\phi} = -\alpha \hat{r}$$

$$\vec{v} = -\alpha \hat{r}$$

